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The Study of Noise Suppression in a Multispectral Imaging System Based on CMOS Image Sensors

Ziheng Shao *

Jinan No.3 Middle School, Jinan, Shandong, China
*Corresponding Author: aaataosu@qq.com

ABSTRACT

This paper focuses on noise suppression in a multispectral imaging joystick system based on CMOS image sensors. CMOS image sensors offer advantages such as low power consumption, high integration, and cost-effectiveness in multispectral imaging systems; however, noise remains a major bottleneck affecting imaging quality. This study provides a detailed analysis of the sources and characteristics of thermal noise, shot noise, and reset noise, and proposes corresponding noise suppression algorithms, including cooling techniques, mean filtering algorithms, and wavelet transform algorithms. Based on experimental design and parameter settings, the effectiveness and processing speed of each algorithm were evaluated. The results demonstrate that optimized noise suppression algorithms significantly improve the signal-to-noise ratio and image quality. Finally, this paper summarizes the research findings and anticipates future research directions, including algorithm optimization.

KEYWORDS

CMOS Image Sensor; Multispectral; Imaging System; signal-to-noise ratio; Noise; Imaging model

1. INTRODUCTION

1.1. Research Background and Significance

"The Study of Noise Suppression in a Multispectral Imaging System Based on CMOS Image Sensors" aims to address the common noise issues encountered in multispectral imaging. As one of the core components of multispectral imaging systems, the noise in CMOS image sensors significantly impacts the imaging results. These noises, including dark current noise and readout noise, reduce the signal-to-noise ratio of the image, affect image quality, and compromise the reliability of the data, thereby affecting the accuracy of subsequent data analysis and application. By researching noise suppression techniques for CMOS image sensors, it is possible to fundamentally resolve the noise problems in multispectral imaging systems.

1.2. Related Work Overview

Multispectral imaging technology is currently widely used, particularly in fields such as agriculture, environmental monitoring, and medical imaging. Traditional imaging systems primarily use CCD sensors; however, CMOS image sensors have gradually become the preferred choice due to their low power consumption, high integration, and cost-effectiveness. Since the advent of CMOS image sensors, noise issues have been a bottleneck limiting the imaging quality of these sensors. The noise problems severely restrict the imaging quality of image sensors and have continuously impacted the development and application of image sensor technology. The overall performance of sensors is

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ultimately limited by the noise of the system itself. Low noise translates to a high dynamic range. Therefore, having a clear understanding and conducting research on the noise and its suppression methods for image sensors can provide better insight into the overall sensor design and imaging mechanisms. This allows for minimizing noise impact during design and application, thereby achieving a high signal-to-noise ratio and a wide dynamic range in image sensor systems [1].

1.3. Research Content

Analyze the sources of noise in CMOS image sensors used in multispectral imaging, including thermal noise, shot noise, reset noise, and others.

2. BASIC KNOWLEDGE INTRODUCTION

2.1. Principle of CMOS Image Sensors

Complementary Metal-Oxide-Semiconductor (CMOS) is a semiconductor technology used in most integrated circuits (ICs) today, also known as chips or microchips. CMOS transistors are based on Metal-Oxide-Semiconductor Field-Effect Transistor (MOSFET) technology [2].

Photoelectric Conversion: Incident photons are absorbed by the photodiode, generating photoelectrons.

Charge Accumulation and Transfer: Photoelectrons are accumulated within the pixel unit and transferred to the readout node via transistors.

Signal Readout: The accumulated charge is converted into a voltage signal, amplified by a source follower transistor, and then read out row by row through the column selection circuit.

Analog-to-Digital Conversion: The analog signal is converted into a digital signal by an ADC (Analog-to-Digital Converter), laying the foundation for subsequent image processing and storage.

2.2. Overview of Multispectral Imaging Technology

Multispectral imaging technology captures spectral information across multiple wavelength ranges to obtain images of target objects. It is widely used in agriculture, environmental monitoring, medical imaging, remote sensing, and military applications.

2.3. Analysis of Noise Types and Characteristics

2.3.1. Thermal Noise

Source: Thermal noise is caused by the thermal motion of electrons within the sensor, primarily occurring in the photodiode and readout circuitry.

Characteristics: Thermal noise is proportional to temperature and typically follows a Gaussian distribution. As the temperature increases, thermal noise significantly increases.

Suppression Methods: Lowering the sensor's operating temperature, using high-quality materials and designs, and employing more efficient cooling systems.

2.3.2. Shot Noise

Source: Shot noise is caused by the randomness in the arrival of photons at the sensor and the photoelectric conversion process. In the photodiode, the number of photoelectrons generated by incident photons is random, leading to shot noise. Due to its nature being determined by the quantum characteristics of photons, shot noise is an unavoidable type of noise.

Characteristics: The magnitude of shot noise is proportional to the signal, making its impact particularly significant under low light conditions. In high light conditions, the proportion of shot noise relative to the signal is smaller. Shot noise is a type of white noise, meaning its statistical properties are the same across all spatial and temporal points, with no fixed pattern.

Suppression Methods: Increasing incident light intensity, extending exposure time, using high-sensitivity sensors, and employing advanced image processing algorithms.

2.3.3. Reset Noise

Source: Reset noise occurs during the reset phase of each pixel unit. When the capacitor in the pixel is reset to the reference voltage by the reset transistor discharging, thermal agitation can leave a certain amount of uncertain charge on the capacitor. The voltage fluctuation caused by this thermal agitation is known as reset noise.

Characteristics: Reset noise is random and related to the pixel's reset operation, being proportional to temperature. In high-temperature environments, reset noise becomes more pronounced. It appears as a fixed spatial noise pattern in images, significantly affecting image quality, especially under low light conditions.

Suppression Methods: Increasing the capacitor value, optimizing the reset circuit design, lowering the operating temperature, and using post-processing algorithms.

3. NOISE CHARACTERISTICS ANALYSIS

3.1. Image Acquisition and Preprocessing

In multispectral imaging systems, noise analysis for image acquisition and preprocessing is fundamental. Effective acquisition and processing can accurately capture noise characteristics, laying the foundation for subsequent research on noise suppression. The following are methods for acquiring images:

3.1.1. Image Acquisition

Multispectral Image Acquisition: Equipment Selection: Use high-performance multispectral cameras equipped with appropriate filters to capture images of the target scene in different spectral bands.

Control Variables: Ensure consistency in parameters such as lighting conditions, exposure time, and aperture settings for each shot to avoid introducing additional noise.

Environmental Control: Conduct image acquisition in a controlled lighting environment to prevent natural light variations from affecting the images.

3.1.2. Image Preprocessing

(1) Bad Pixel Removal

Bad Pixel Detection: Identify the locations of bad pixels by analyzing abnormal pixel values in the image. Bad pixels typically appear as constant high or low values.

Bad Pixel Correction: Use interpolation algorithms to repair bad pixels, replacing them with the average value of nearby normal pixels.

(2) Denoising Preprocessing

Filtering: Apply spatial filtering or frequency domain filtering (such as wavelet transform) to remove random noise from the image.

Advanced Denoising Algorithms: Use learning models to eliminate complex noise patterns, enhancing image clarity and signal-to-noise ratio.

4. NOISE SUPPRESSION ALGORITHM DESIGN AND OPTIMIZATION

4.1. Thermal Noise

4.1.1. Algorithm Design

Cooling Techniques: Refrigeration Equipment: Use devices such as semiconductor coolers or liquid nitrogen cooling to lower the sensor's operating temperature, thereby reducing the generation of thermal noise at the source.

Temporal Averaging: Multi-frame Averaging: Perform averaging on images taken from multiple exposures to reduce the impact of random thermal noise.

Filtering Algorithms: Gaussian Filtering: Use a Gaussian filter to smooth the image and suppress thermal noise.

Median Filtering: Apply a median filter to remove noise from the image.

4.1.2. Algorithm Optimization

Parameter Optimization: Filter Window Size: Optimize the filter window size through experiments or automatic optimization algorithms to achieve the best noise reduction effect.

Multiscale Processing: Multiscale Filtering: Apply Gaussian and median filtering at different scales to enhance the ability to identify and process thermal noise.

4.2. Shot Noise

4.2.1. Algorithm Design

(1) Mean Filtering Algorithm

Mean Filter: Use a mean filter to smooth the image and suppress shot noise. The mean filter reduces noise by averaging the pixel values within the filter window [3].

Filter Window Size: Choose an appropriate window size that can smooth out noise without losing image details.

(2) Increasing Incident Light Intensity

Increase Illumination Intensity: Enhance the brightness of the light source to increase the incident light intensity and reduce the ratio of shot noise to the signal. Since photocurrent shot noise is related to illumination intensity, it is difficult to eliminate it without reducing the sensor's quantum efficiency.

(3) Extending Exposure Time

Appropriate Extension of Exposure Time: Increase exposure time to accumulate more photons, enhancing signal strength and reducing the impact of shot noise on the image.

4.2.2. Algorithm Optimization

(1) Multiscale Processing

Multiscale Mean Filtering: Apply mean filtering at different scales to improve the suppression capability for shot noise.

(2) Adaptive Mean Filtering

Adjust the filter window size and weights adaptively based on the local characteristics of the image to enhance denoising effectiveness.

4.3. Reset Noise

4.3.1. Algorithm Design

(1) Wavelet Transform Algorithm

Wavelet Denoising: Use wavelet transform to decompose the image signal into different scales and suppress noise in the wavelet domain [4, 5]. The steps of wavelet denoising include wavelet decomposition, thresholding, and wavelet reconstruction.

(2) Sensor Design

Bus Capacitance Control: Pay attention to the control of bus capacitance during sensor design, which can effectively suppress reset noise [6].

4.3.2. Algorithm Optimization

(1) Parameter Optimization

Threshold Optimization: Optimize the threshold parameters for wavelet denoising through experiments or automatic optimization algorithms to achieve the best noise reduction effect.

(2) Multiscale Processing

Multiscale Wavelet Denoising: Apply wavelet denoising at different scales to enhance the ability to identify and process reset noise.

5. ALGORITHM VERIFICATION AND EVALUATION

5.1. Experimental design and parameter setting

5.1.1. Dataset Selection

Standard Datasets: Use publicly available multispectral image datasets, which include various scenes and noise types, to facilitate general validation.

Custom Datasets: Collect multispectral images based on actual application scenarios to create a dataset that includes real noise characteristics.

5.1.2. Experimental Environment:

Hardware Environment: Equip hardware similar to that used in actual applications.

Software Environment: Create an experimental platform that includes the suppression algorithms to ensure the algorithms can be compared under the same conditions.

5.1.3. Parameter Settings

Filter Parameters: Set appropriate filter parameters based on different noise types.

Experimental Conditions: Conduct experiments under varying conditions of illumination, temperature, etc., to seek general rules.

5.2. Algorithm Suppression Effect Evaluation

5.2.1. Subjective Evaluation

Visual Effect: Observe the denoised images to assess their visual effect and quality improvement.

Expert Evaluation: Invite domain experts to subjectively evaluate the denoising effect and provide professional opinions.

5.2.2. Objective Evaluation

Peak Signal-to-Noise Ratio (PSNR): Measure the similarity between the denoised image and the original image.

Structural Similarity Index (SSIM): Assess the similarity between the denoised image and the original image [7, 8].

Signal-to-Noise Ratio (SNR): Evaluate the improvement in the signal-to-noise ratio before and after denoising.

5.3. Algorithm Processing Speed Testing

5.3.1. Execution Time Measurement

Single Image Processing Time: Record the denoising time for each image to evaluate real-time performance.

Batch Processing Time: Measure the time taken to process a large number of images in batch mode to assess the algorithm's efficiency.

5.3.2. Performance Comparison

Benchmark Algorithm Comparison: Select classical noise suppression algorithms as benchmarks for processing speed comparison.

Hardware Acceleration: Evaluate processing speed in different hardware environments to analyze the impact of hardware upgrades on algorithm performance.

5.3.3. Algorithm Complexity Analysis

Time Complexity: Analyze the time complexity of the algorithm to evaluate its processing speed on datasets of different scales [9].

Space Complexity: Assess the space complexity of the algorithm to understand its hardware requirements, such as memory usage.

6. ANALYSIS AND SUMMARY

6.1. Experimental Results Analysis

6.1.1. Thermal Noise Suppression Effect

Cooling Techniques: Experimental results indicate that by lowering the sensor's operating temperature, thermal noise is significantly reduced.

Temporal Averaging: Multi-frame averaging effectively reduces thermal noise, resulting in a notable increase in the signal-to-noise ratio (SNR) of the images.

Filtering Algorithms: Both Gaussian filtering and median filtering effectively smooth out thermal noise.

6.1.2. Shot Noise Suppression Effect

Mean Filtering Algorithm: The mean filter performs well in suppressing shot noise. Visually, the mean filter reduces shot noise in the images while preserving image details.

Increasing Incident Light Intensity and Extending Exposure Time: By increasing light source brightness and extending exposure time, the noise-to-signal ratio is significantly reduced, leading to noticeable improvements in image quality.

6.1.3. Reset Noise Suppression Effect

Wavelet Transform Algorithm: Wavelet denoising shows excellent performance in suppressing reset noise. The visual quality of the images is significantly improved, effectively reducing noise while maintaining image quality and details.

6.2. Paper Summary and Future Prospects

This paper addresses the noise issues (thermal noise, shot noise, and reset noise) in CMOS image sensors and proposes a series of suppression methods, summarized as follows:

(1) Thermal Noise: Effective suppression of thermal noise was achieved through cooling techniques, temporal averaging, and filtering algorithms, thereby improving image quality.

(2) Shot Noise: The mean filtering algorithm demonstrated excellent performance in suppressing shot noise, significantly increasing the image's signal-to-noise ratio, as shown by experimental results.

(3) Reset Noise: The wavelet transform algorithm effectively suppressed reset noise, resulting in noticeable improvements in image quality.

6.3. Future Research Directions

(1) Algorithm Optimization: Further optimize the parameter settings and processing flow of noise suppression algorithms to enhance their reliability and applicability.

(2) Hardware Acceleration: Explore more hardware acceleration methods to further improve the real-time processing capability of the algorithms.

(3) New Multispectral Imaging Technologies: Develop new noise suppression algorithms by integrating advanced technologies such as deep learning.

(4) Practical Applications: Conduct more testing and refinements in real-world application scenarios to ensure the algorithms' effectiveness in various complex environments.

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